

Machine learning to characterize antimicrobial resistance

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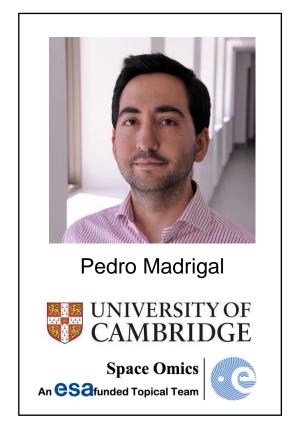
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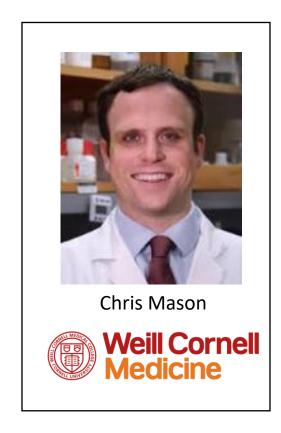
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Acknowledgements



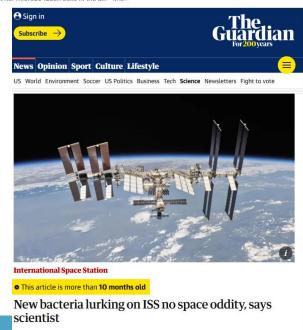








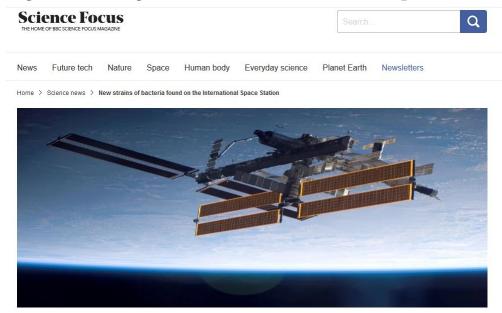
International Space Station astronaut Karen Nyberg made a video showing how she washed her hair-and inevitably put her microbe-laden cells in the air. NASA



New species were discovered in the International Space Station -

but they probably didn't come from outer space

Why Study Microbes in Space?



New strains of bacteria found on the International Space Station



ARTICLE

Space Station conditions are selective but do not alter microbial characteristics relevant to human health

Maximilian Mora¹, Lisa Wink¹, Ines Kögler¹, Alexander Mahnert¹, Petra Rettberg[©] ², Petra Schwendner³, René Demets⁴, Charles Cockello ³, Tatiana Alekhova⁵, Andreas Klingl⁶, Robert Krause^{1,7}, Anna Zolotariofo ³, Alina Alexandrova⁵ & Christine Moissl-Eichinger^{1,7}

https://www.nature.com/articles/s41467-019-11682-z.pdf



Previously unknown bacteria discovered on the space station could help grow plants





Would you eat food grown on Mars? 01:05



Jan 22, 2020





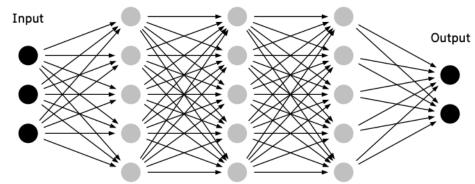
Study finds space station microbes are no more harmful than those found in similar ground environments

Rest assured, microbes do not, it turns out, become "super bugs" in space.



Supervised Machine Learning for Prediction of Antimicrobial Resistance

- A Deep Learning approach for predicting antibiotic resistance genes (ARGs) from metagenomic data
- Curated database of 14,933 genes from three databases (CARD, ARDB, and UNIPROT)
- Model:
 - Input layer of 4333 units that correspond to the ARGs from ARDB and CARD
 - 4 hidden layers of 2k, 1k, 500 and 100 units
 - Output layer of DNN consists of 30 units that correspond to antibiotic resistance categories (102 antibiotics consolidated into 30 antibiotic categories)
- Two models: SS (short read sequences) and LS (full gene length sequences)
 - The CARD and ARDB genes were selected to represent known ARGs, whereas the UNIPROT genes were used for training and validation of the models.



A simple model of Deep Feedforward Neural Network Structure (dotnets/Graphviz)

DeepARG, Arango-Argoty et al. Microbiome (2018)

Antibiotic Resistance Mechanisms

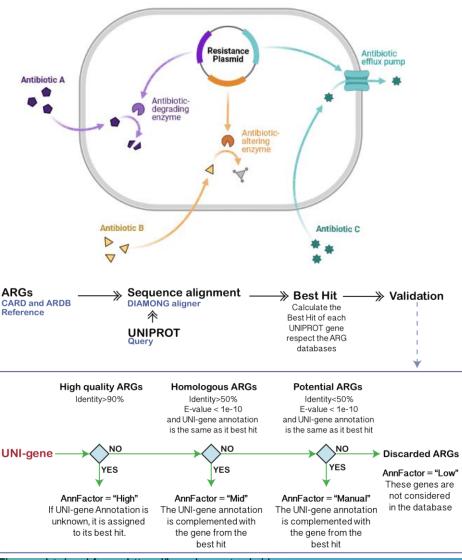
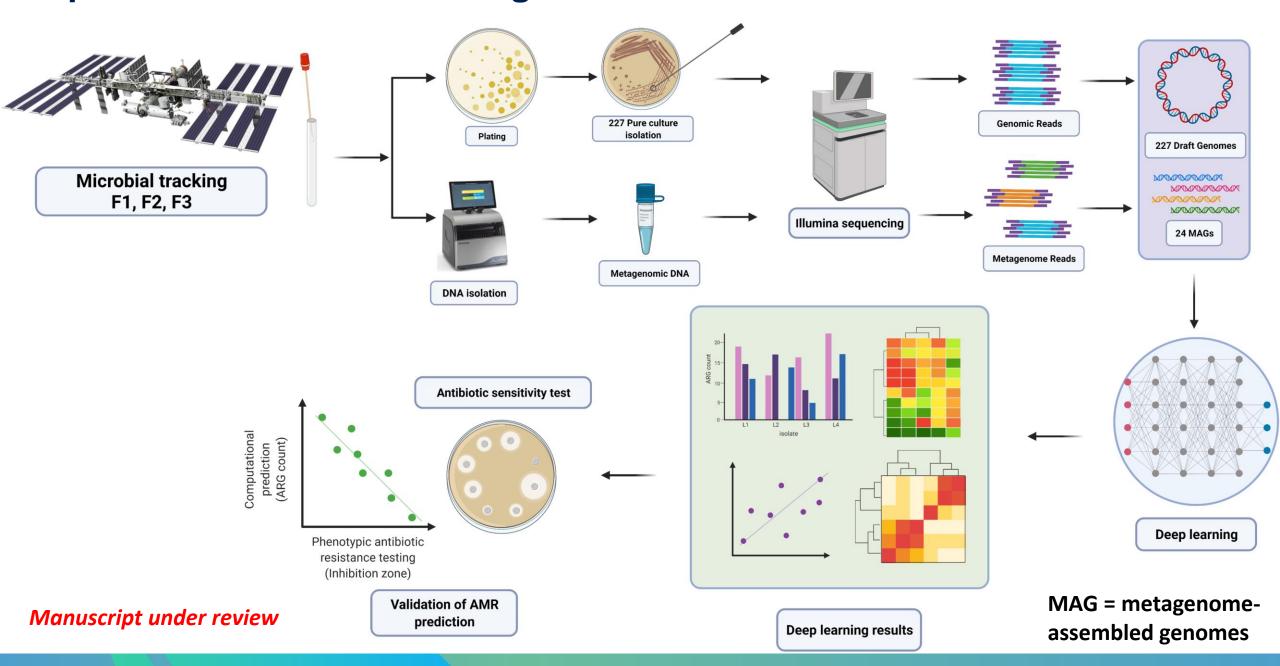
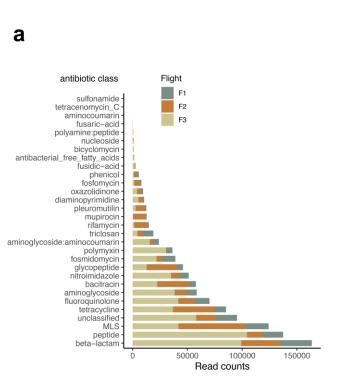


Figure obtained from: https://bench.cs.vt.edu/deeparg

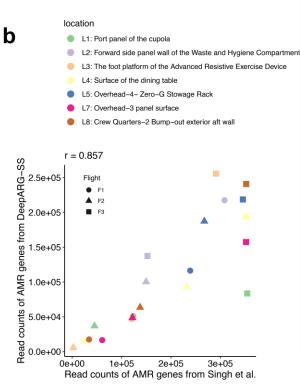
Supervised Machine Learning for Prediction of Antimicrobial Resistance



Prediction of Antibiotic Resistance Genes (ARGs) using a pre-trained DeepARG-SS model

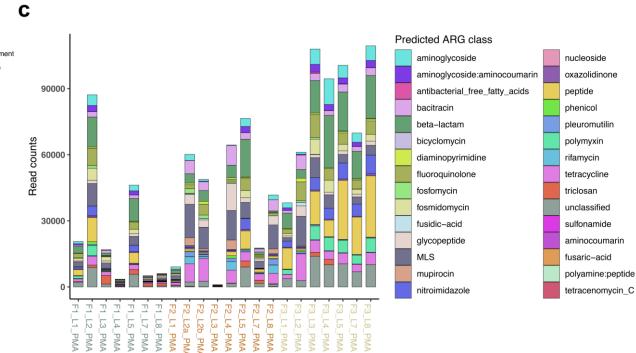


Distribution of ARG read counts across antibiotic classes for the three flights



Correlation of read counts found by DeepARG-SS and those in Singh et al. (2018).

Propidium monoazide (PMA) is a photoreactive DNA-binding dye that



Read counts of ARG class across flights for each location for propidium monoazide (PMA)treated samples in Singh et al. (2018)

> Singh et al. Microbiome https://doi.org/10.1186/s40168-018-0585-2

Microbiome

RESEARCH

Open Access

Succession and persistence of microbial

communities and antimicrobial resistance genes associated with International Space

Station environmental surfaces

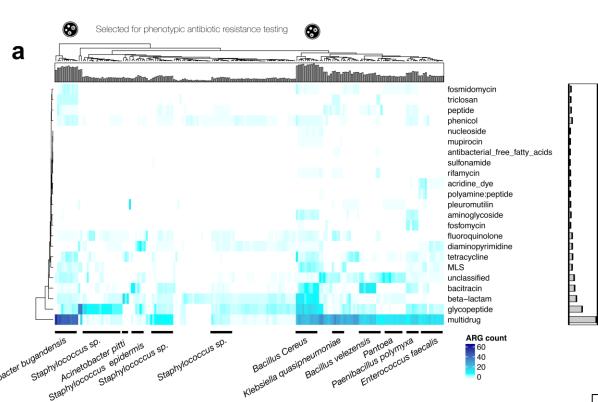
preferentially binds to dsDNA. It is used to detect viable microorganisms by qPCR

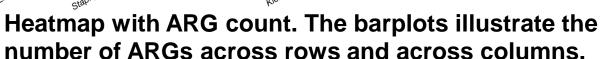
ARGs detected in Open Reading Frames (ORFs) in metagenome-assembled genomes (MAGs) from **PMA-treated samples** Total number of ARGs predicted a 100 for each flight and location Distribution of ORFs location **DeepARG** classification location S₁P L1: Port panel of the cupola _ probability and best-hit Flight S2P predicted L2: Forward side panel wall of the Waste and Hygiene Compartment F1 S3P identity in MAGs L3: The foot platform of the Advanced Resistive Exercise Device ▲ F2 S4P retrieved from the ISS ■ F3 genes L4: Surface of the dining table 0.90 S7P L5: Overhead-4- Zero-G Stowage Rack S8P L7: Overhead-3 panel surface L8: Crew Quarters-2 Bump-out exterior aft wall 0.85 Important! Flight identity low sequence identity (mostly discarded by traditional approaches) but C high probability MAGs with at least 1 predicted ARG Predicted ARG class 3F S8P Kalamiella piersonii Kalamiella piersonii antibacterial free fatty acids 3F S7P Kalamiella piersonii beta-lactam https://www.mdpi.com/2076-0817/9/9/711ntoea dispersa glycopeptide Number of ARGs triclosan 3F S3P Klebsiella pneumoniae bacitracin S2P_Staphylococcus saprophyticus predicted for each MAG. phenicol 3F_S1P_Kalamiella piersonii tetracycline S8P_Paenibacillus polymyxa Most common antibiotic unclassified S7P_Sphingomonas sanguinis location diaminopyrimidine 2F_S7P_Methylobacterium class (multi-drug) not S5P Pantoea brenneri fosmidomycin shown 2F_S5P_Paenibacillus polymyxa MLS fluoroquinolone S2P Staphylococcus saprophyticus mupirocin 2F S1P Acinetobacter pittii pleuromutilin S5P Pantoea brenneri aminoglycoside 1F S2P Enterobacter bugandensis peptide Manuscript under review 1F S1P Klebsiella polymyxin ARGs predicted in ORFs

F1 F2 F3

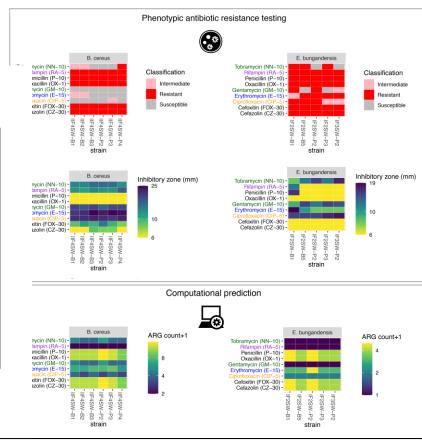
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Heatmap and clustering of ARG counts detected in Microbial Tracking-1 pure strains isolated from the ISS and Antibacterial Susceptibility Tests (AST) validations

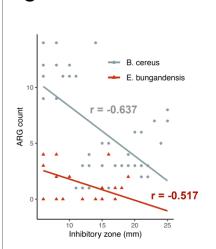




Manuscript under review



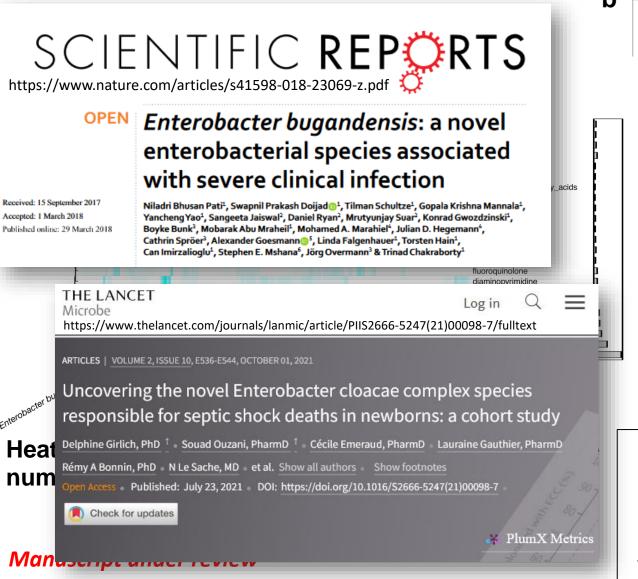
Antibacterial Susceptibility Tests
(AST) on *E. bugandensis* and *B*cereus strains for several
antibiotics (top), and comparison
with machine learning predictions
shown in (a)

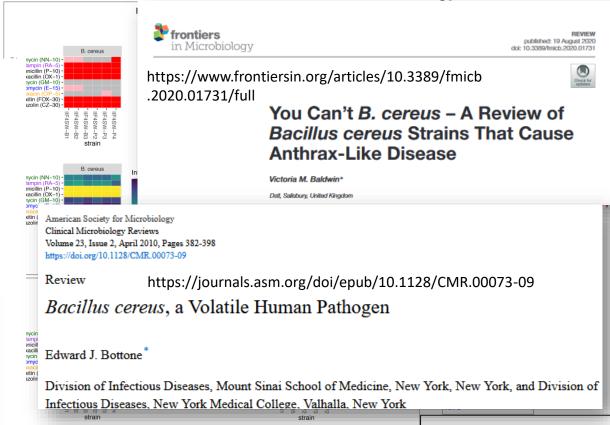




Scatterplot of zone of inhibition value (in mm.) and ARG count shown in (b), together with a linear model fit.

Heatmap and clustering of ARG counts detected in Microbial Tracking-1 pure strains isolated from the ISS and Antibacterial Susceptibility Tests (AST) validations





Antibacterial Susceptibility Tests
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